Data Visualisation - GGPlot2

Webbinar - corona edition - Thomas Lin Pedersen from R Grammar of Graphics (ggplot2)

Data import

readr - reading tabulated data readxl - reading data from excel files haven - reading data from spss

Data Manipulation

tidyr - tidyverse dplyr - tidyverse data.table

The Grammar of Graphics

Data is not just data Representation defines what can be done with it Grammar requires a tidy format

Data: Obviously

Aesthetic Mappings: this column represents x-axis, y-axis, etc **Facet** mapping: link variables in the data to panels in the facet layout

Statistics: Applying some sort of transformation to the data to be able to plot it correctly, ex. Bar graphs

need specified count, so data must contain that count.

Scales: graphical representable property, color, shapes, ex. mercedes -> blue, toyota -> red, etc.

Scales takes into consideration the different types of data

Geometries: how to specify how to interpret the data, ex. geo_point will plot as a point while geo_line will plot as a line (connect the points)

Facets: Splitting the data into different subplots, small multiples of the same data.

Coordinates: Defines the physical mapping of the aesthetics to the paper.

Theme: color of the background, style, pure style.

Importing & Loading with Pacman

pacman:: p_load(pacman, dplyr, GGally, ggplot2, ggrepel, patchwork, gifski, ggforce, ggthemes, maps, sf

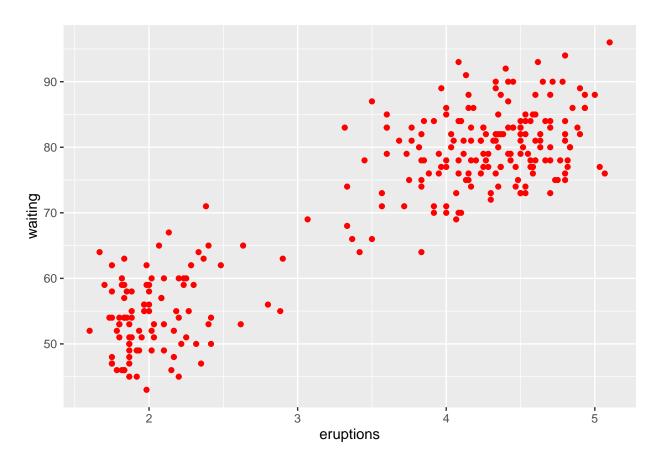
Simple Scatter in GGPlot

Using the example data faithful, the simplest arguments for making a plot are the data, the columns (aesthetics), and the way to read it (geometries)

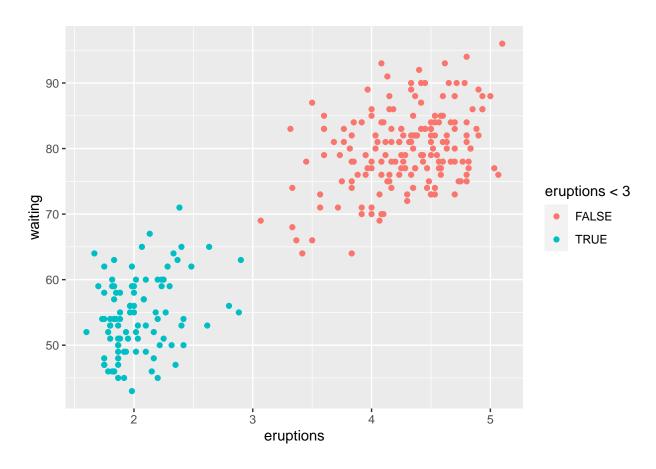
head(faithful)

```
eruptions waiting
##
## 1
         3.600
                     79
## 2
         1.800
                     54
## 3
         3.333
                     74
         2.283
                     62
## 4
## 5
         4.533
                     85
                     55
## 6
         2.883
```

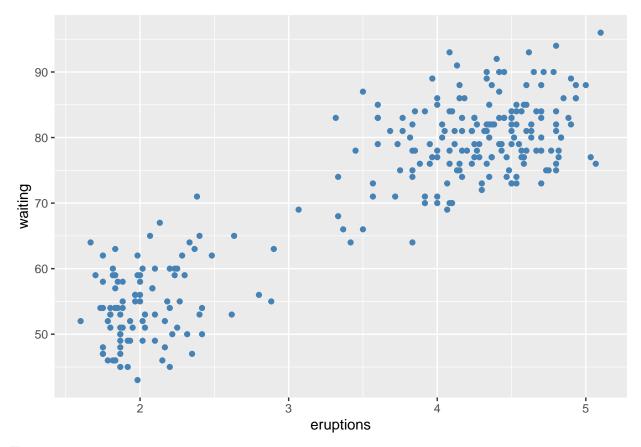
```
ggplot(faithful, aes(eruptions, waiting)) +
geom_point(color='red')
```



```
ggplot(faithful) +
  geom_point(aes(eruptions, waiting, colour = eruptions < 3))</pre>
```

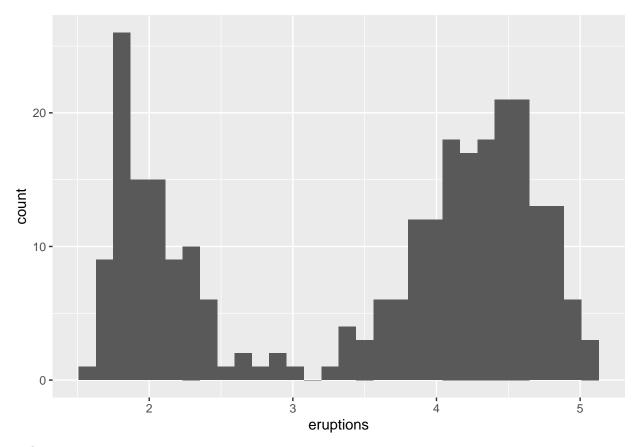


```
ggplot(faithful) +
  geom_point(aes(eruptions, waiting),colour = "steelblue")
```



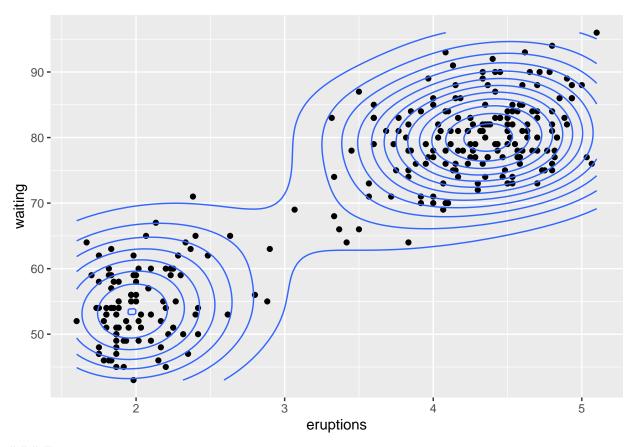
Histogram

```
ggplot(faithful) +
  geom_histogram(aes(eruptions))
```



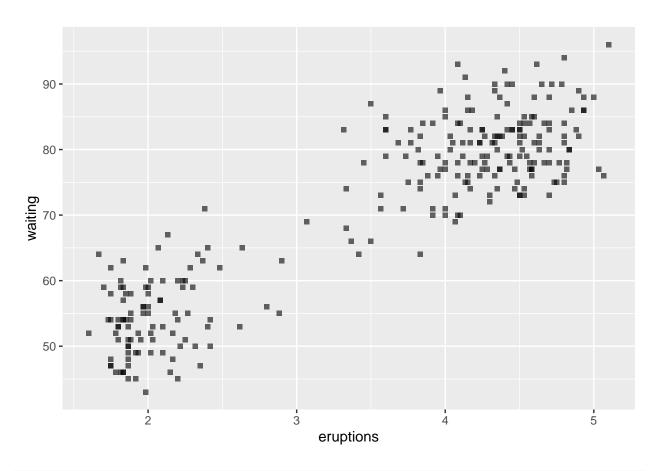
combo

```
ggplot(faithful) + aes(eruptions, waiting) +
  geom_point() +
  geom_density2d()
```

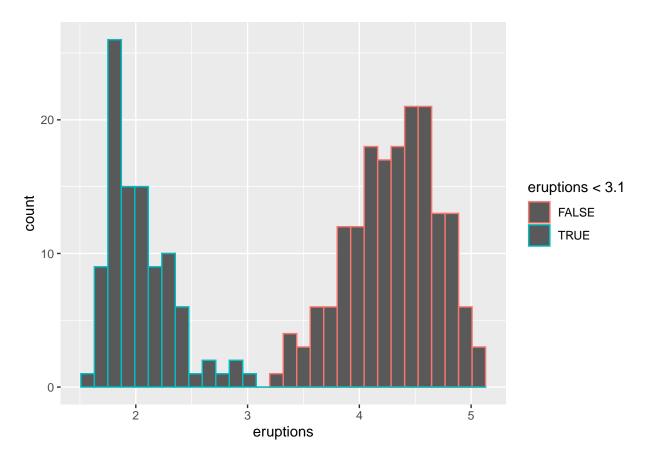


Exercises

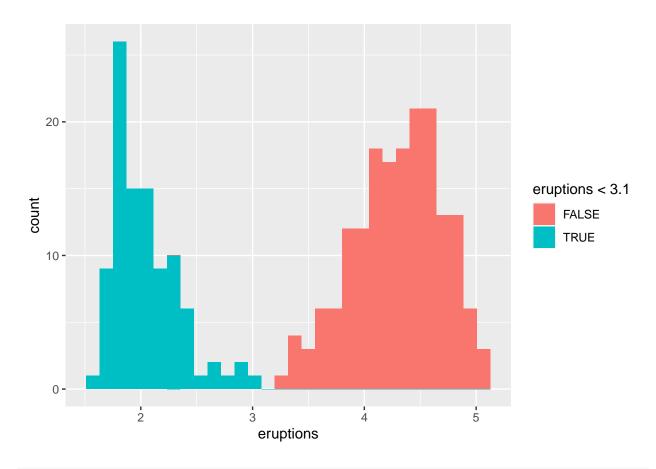
```
ggplot(faithful) +
  geom_point(aes(eruptions, waiting), shape = 'square', alpha = 0.6)
```



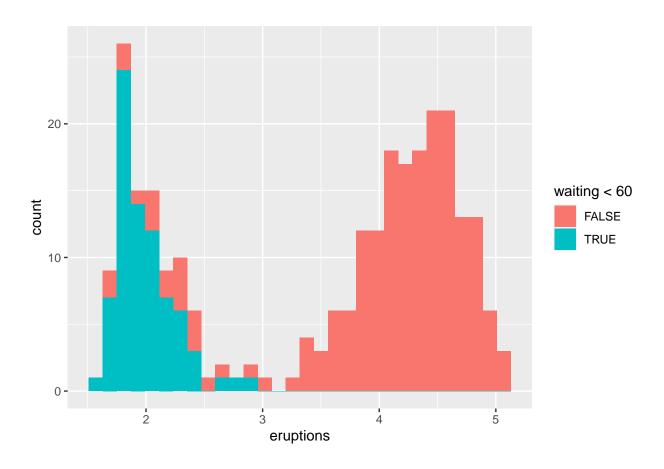
```
ggplot(faithful) +
  geom_histogram(aes(eruptions, colour = eruptions < 3.1))</pre>
```



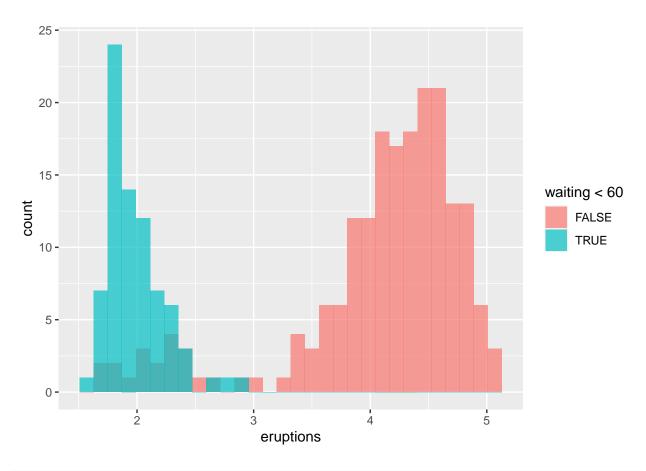
```
ggplot(faithful) +
  geom_histogram(aes(eruptions, fill = eruptions < 3.1))</pre>
```



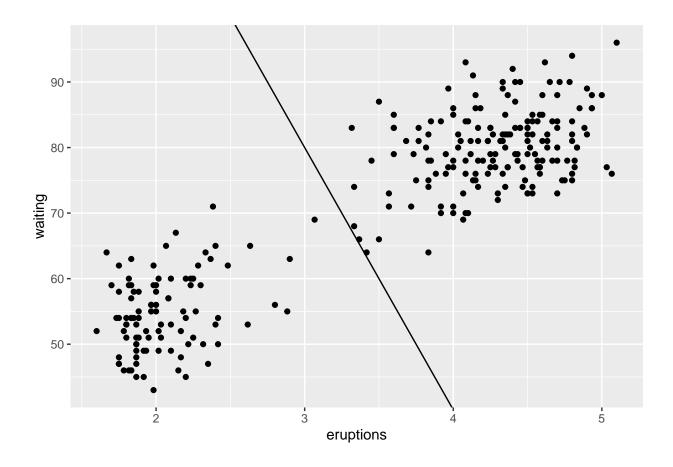
ggplot(faithful) +
 geom_histogram(aes(eruptions, fill = waiting < 60)) #Hard to see, but they're stacked together</pre>



```
ggplot(faithful) +
geom_histogram(aes(eruptions, fill = waiting < 60), position = 'identity', alpha = 0.7) #Position adj</pre>
```



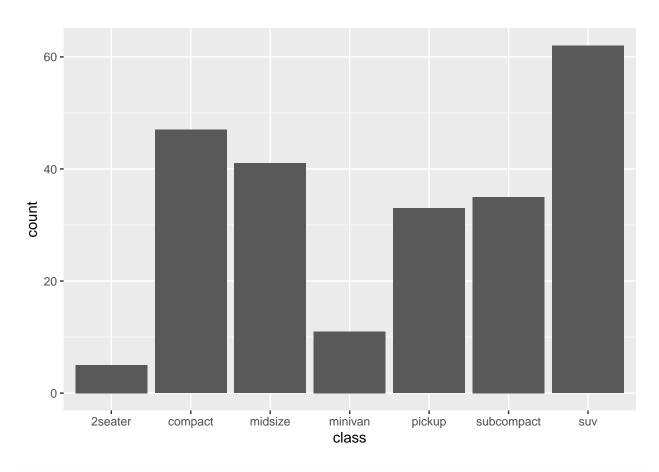
```
ggplot(faithful) +
  geom_point(aes(eruptions, waiting)) +
  geom_abline(slope = -40, intercept = 200)
```



Stats

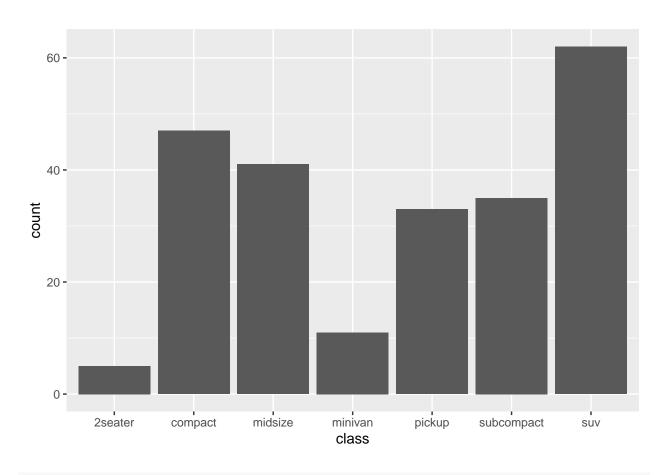
Here geom_bar() uses stat_count() by default to count all the things inside the data. This is all done in the computer, so the computer loads the data and manipulates it. This is not feasible for enormous amounts of data, so its better to tell the data base to do the calculation and safe it on a different data base.

```
ggplot(mpg) +
  geom_bar(aes(class))
```

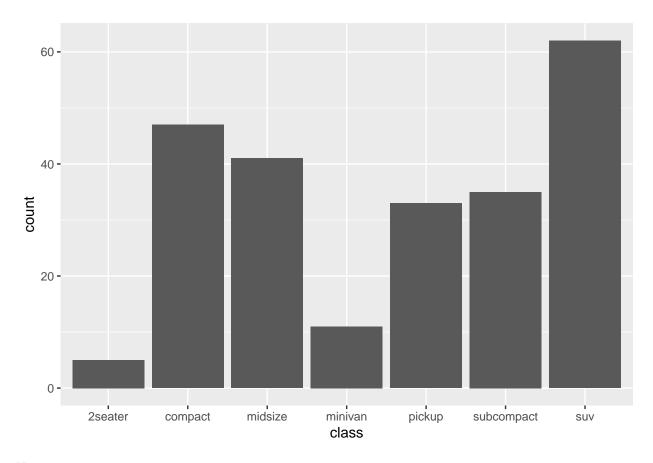


```
mpg_counted <- mpg %>%
  count(class, name = 'count')

ggplot(mpg_counted) +
  geom_bar(aes(class, count), stat = 'identity')
```

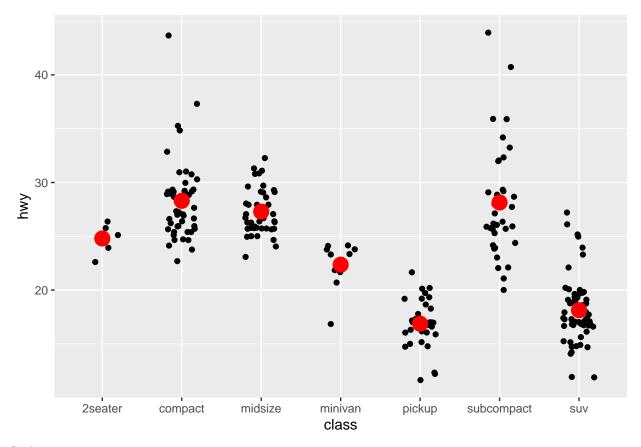


ggplot(mpg_counted) +
 geom_col(aes(class, count)) #geom_col() is a shortcut to avoid putting 'identity' manually



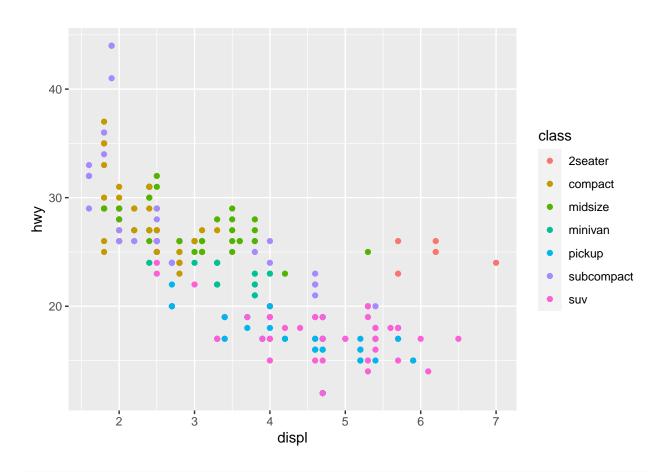
Using stats:

```
ggplot(mpg) +
  geom_jitter(aes(class, hwy), width = 0.2) +
  stat_summary(aes(class, hwy), fun = mean, geom = 'point', colour = 'red', size = 5) # adding a point
```

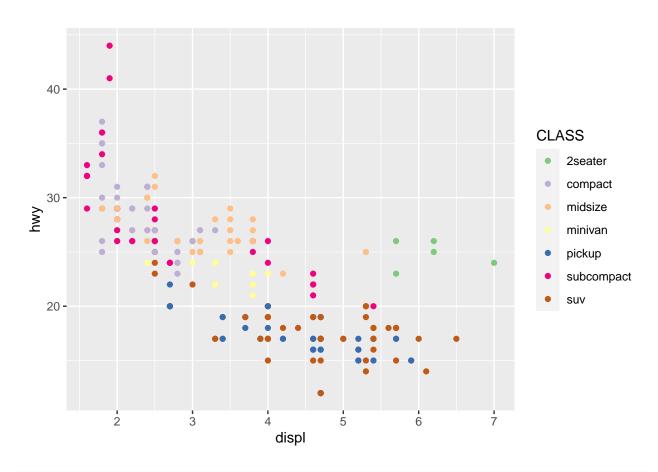


Scales:

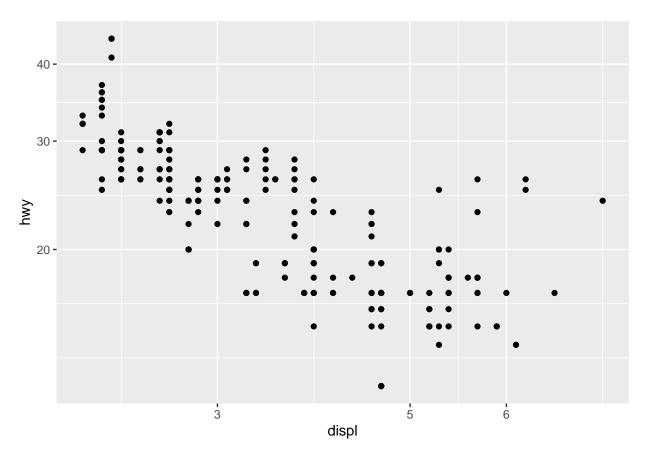
```
ggplot(mpg) +
  geom_point(aes(displ, hwy, colour = class))
```



```
ggplot(mpg) +
  geom_point(aes(displ, hwy, colour = class)) +
  scale_colour_brewer(name = 'CLASS', type = 'qual') #changing the name of the legend and inputing man
```

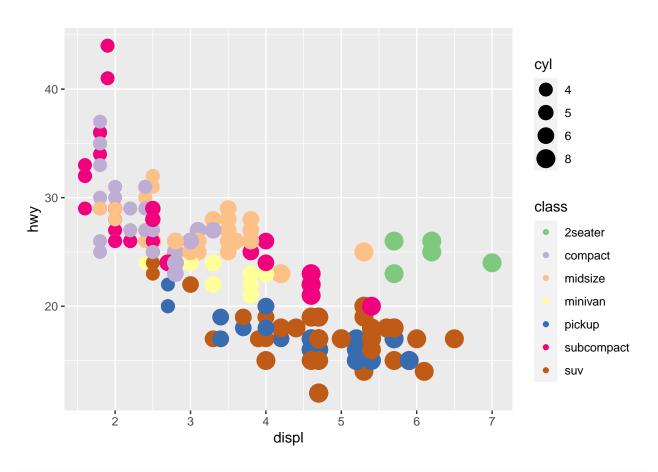


```
ggplot(mpg) +
  geom_point(aes(displ, hwy)) +
  scale_x_continuous(breaks = c(3,5,6))+ #breaks means which numbers you want displayed in the axis.
  scale_y_continuous(trans = 'log10') # trans = transformation
```

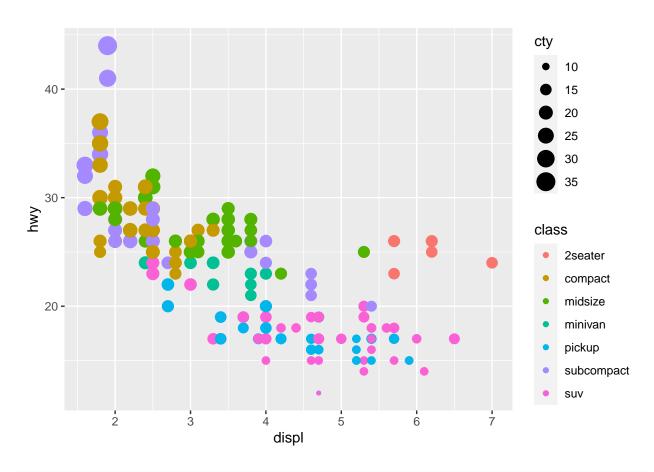


```
#all the display colours
#RColorBrewer::display.brewer.all()

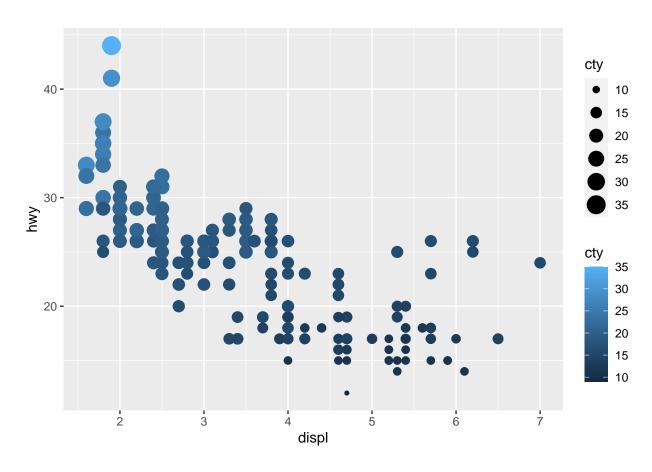
ggplot(mpg) +
  geom_point(aes(displ, hwy, colour = class, size = cyl)) +
  scale_color_brewer(type = 'qual') +
  scale_size_area(breaks = c(4,5,6,8)) #Size is also mapped to a continuous variable only 4,5,6,8
```



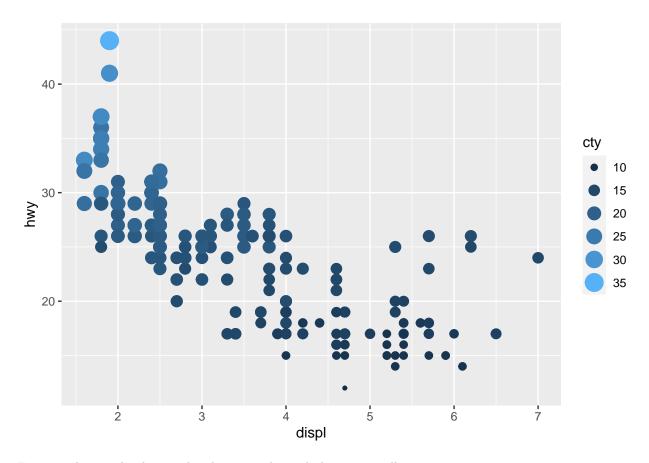
```
ggplot(mpg) +
geom_point(aes(x = displ, y = hwy, colour = class, size = cty))
```



```
ggplot(mpg) +
geom_point(aes(x = displ, y = hwy, colour = cty, size = cty))
```

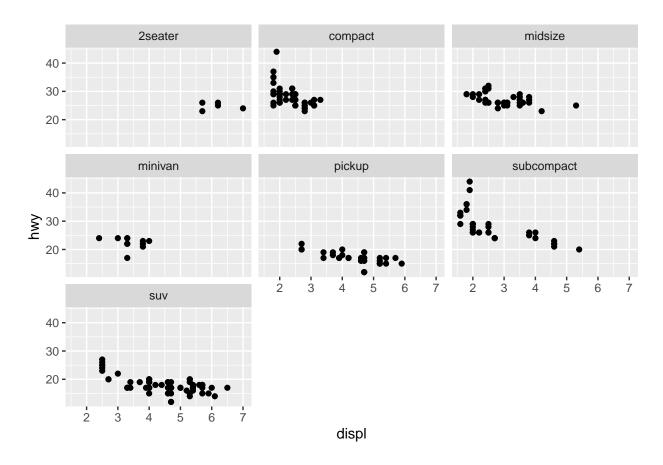


```
ggplot(mpg) +
geom_point(aes(x = displ, y = hwy, colour = cty, size = cty)) +
guides(colour = 'legend') #only merges guides if they're exactly the same!
```

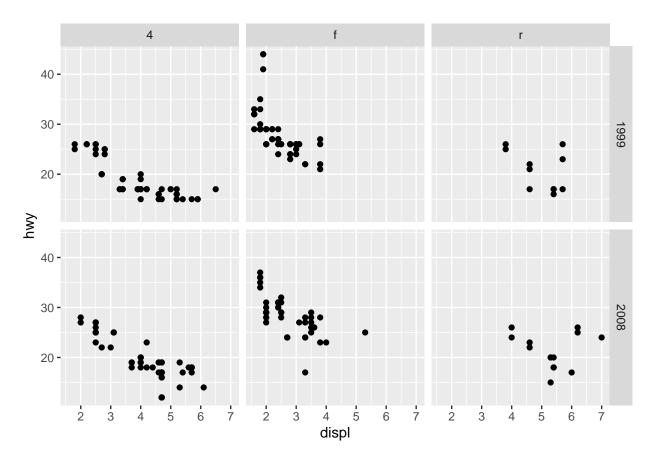


Facets: splitting the data under the same plot, subplots essentially.

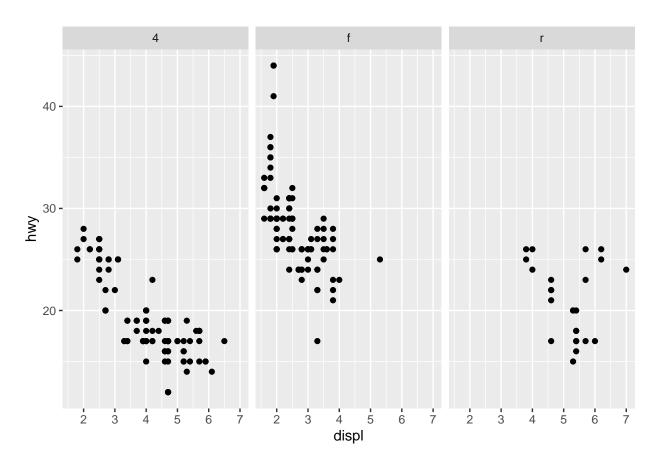
```
ggplot(mpg) +
  geom_point(aes(displ, hwy)) +
  facet_wrap(~ class) #single variable
```



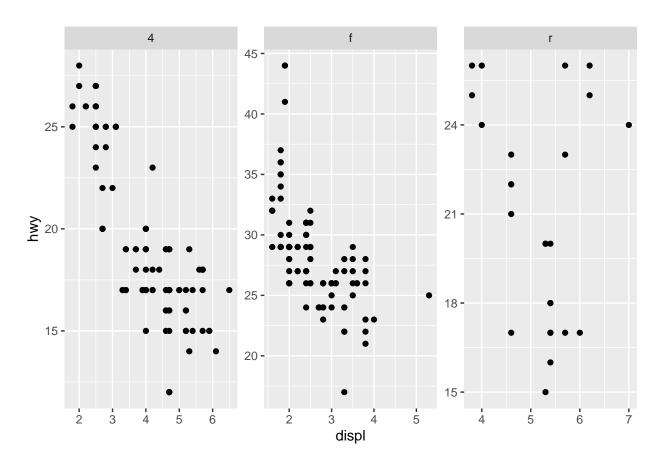
```
ggplot(mpg) +
  geom_point(aes(displ, hwy)) +
  facet_grid(year ~ drv) # two variablies
```



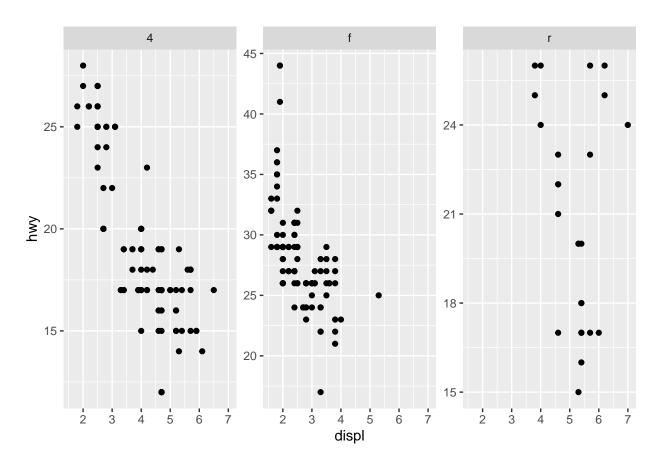
```
ggplot(mpg) +
geom_point(aes(x = displ, y = hwy)) + # same scale for all
facet_wrap(~ drv)
```



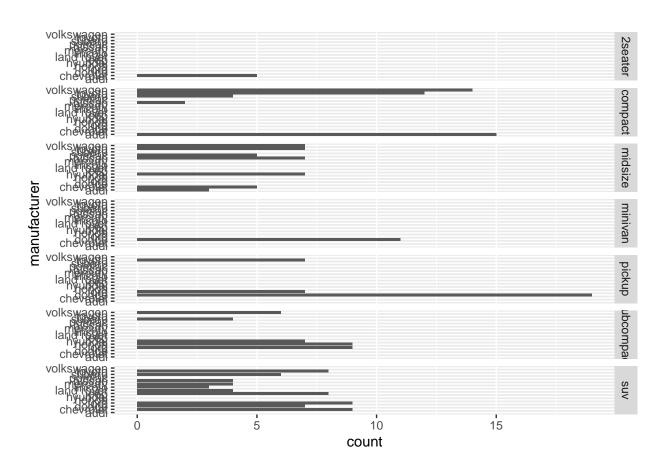
```
ggplot(mpg) +
geom_point(aes(x = displ, y = hwy)) + #different scales for all
facet_wrap(~ drv, scales = 'free')
```



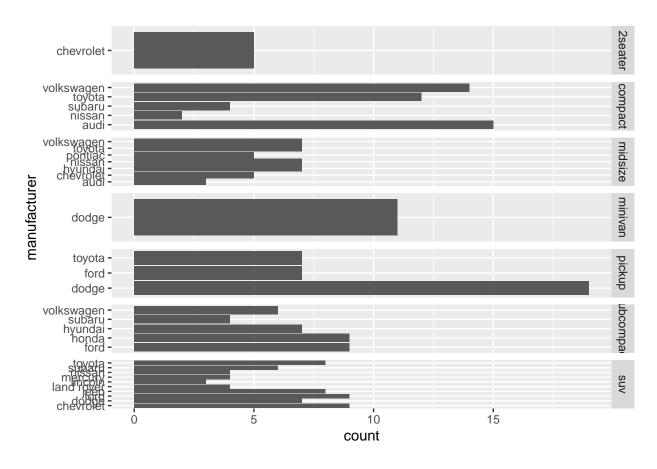
```
ggplot(mpg) +
geom_point(aes(x = displ, y = hwy)) + #only different y scales but same x scale
facet_wrap(~ drv, scales = 'free_y')
```



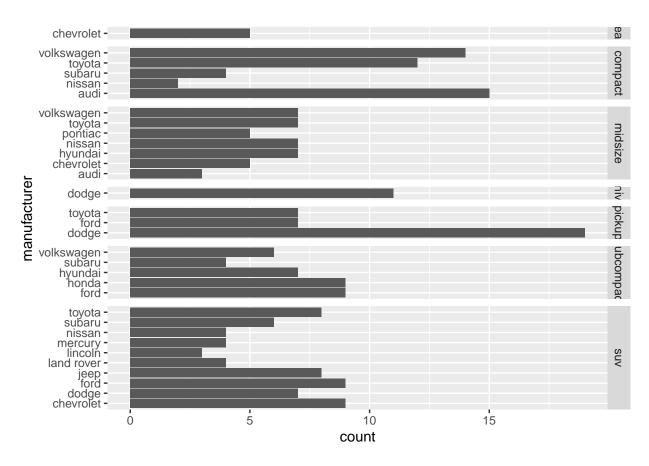
```
ggplot(mpg) +
  geom_bar(aes(y = manufacturer)) +
  facet_grid(class ~ .)
```



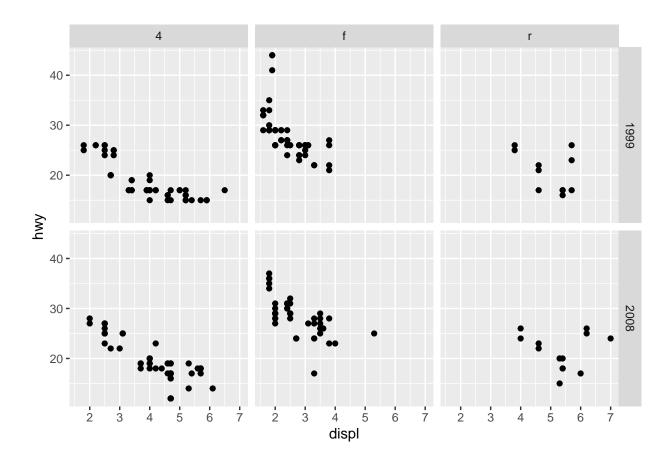
```
ggplot(mpg) +
  geom_bar(aes(y = manufacturer)) +
  facet_grid(class ~ ., scale = 'free_y')
```



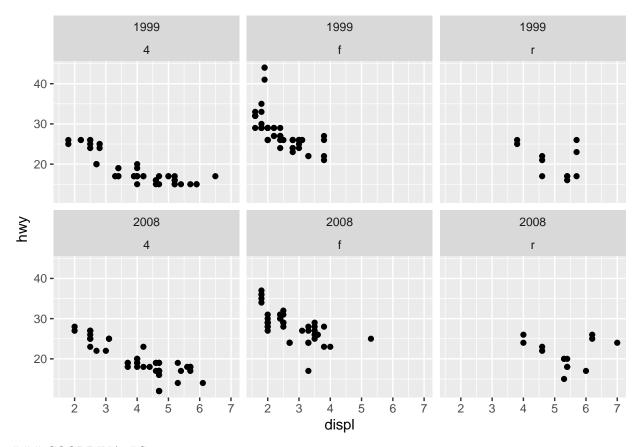
```
ggplot(mpg) +
  geom_bar(aes(y = manufacturer)) +
  facet_grid(class ~ ., space = 'free_y', scale = 'free_y')
```



```
ggplot(mpg) +
  geom_point(aes(x = displ, y = hwy)) +
  facet_grid(year ~ drv)
```

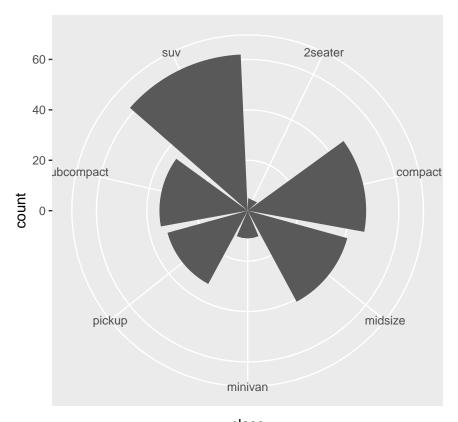


ggplot(mpg) +
 geom_point(aes(x = displ, y = hwy)) + #Less affective way of using it but sure here it is
 facet_wrap(~ year + drv)



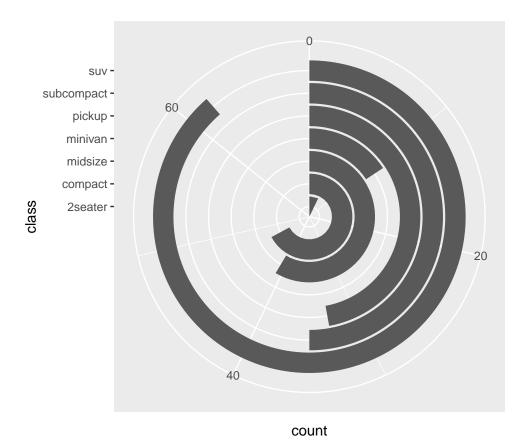
COORDINATES

```
ggplot(mpg) +
  geom_bar(aes(class)) +
  coord_polar()
```



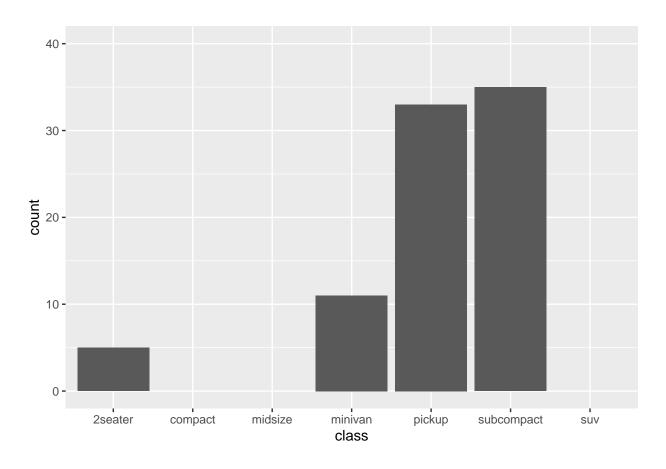
class

```
ggplot(mpg) +
  geom_bar(aes(class)) +
  coord_polar(theta = 'y') +
  expand_limits(y = 70)
```

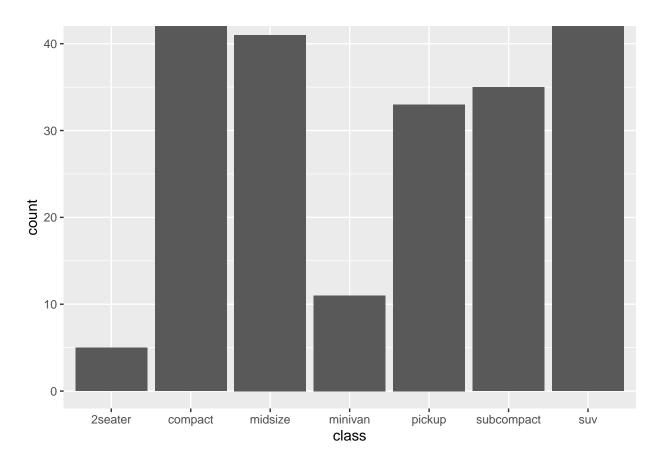


```
ggplot(mpg) +
  geom_bar(aes(class)) +
  scale_y_continuous(limits = c(0,40)) #you're trying to zoom in to your plot but if you try to set the
```

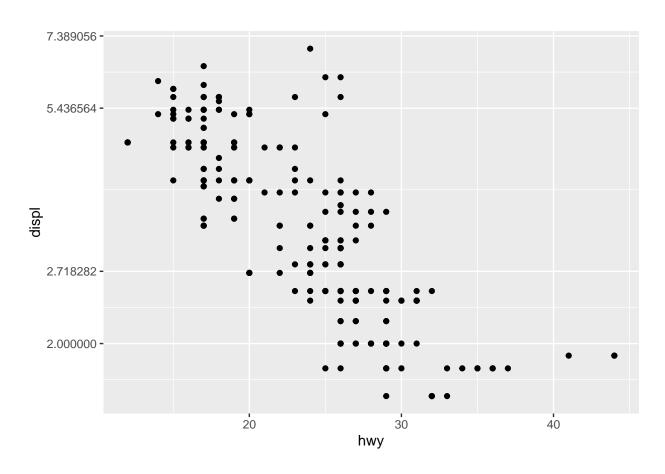
Warning: Removed 3 rows containing missing values (geom_bar).



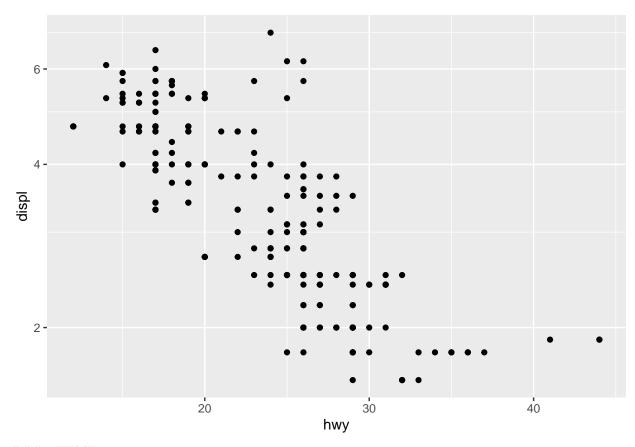
```
\label{eq:cond_model} $\operatorname{ggplot(mpg)} + \\ \operatorname{geom\_bar(aes(class))} + \#that \ is \ why \ you \ set \ the \ limits \ in \ the \ coordinate \ system \ instead, \ doesnt \ mess \\ \operatorname{coord\_cartesian(ylim = c(0,40))}
```



```
ggplot(mpg) +
  geom_point(aes(hwy, displ)) +
  scale_y_continuous(trans = 'log') #breaks are horrible, because i'm doing it in scale
```

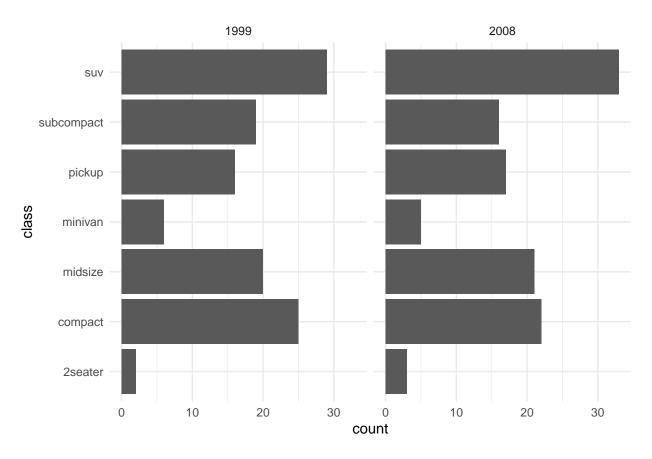


```
ggplot(mpg) +
  geom_point(aes(hwy, displ)) +
  coord_trans(y = 'log') #This only stretches the fabric after the calculations have been done to crea
```



THEME

```
ggplot(mpg) +
  geom_bar(aes(y = class)) +
  facet_wrap(~year) +
  theme_minimal()
```



Number of car models per class

